

## General Compensation Techniques

Linear state feedback has been expressed as:

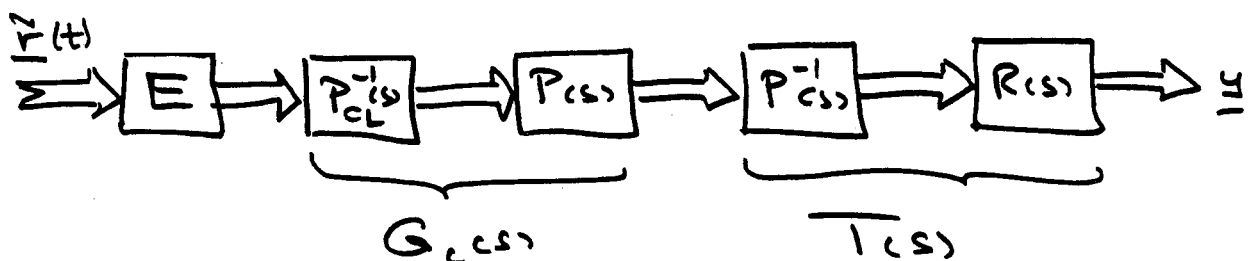
$$\begin{aligned} G_{tot}(s) &= R(s) \cdot P_{CL}^{-1}(s) \cdot E \\ &= T_{tot}(s) \cdot E \end{aligned}$$

We can add  $E$  also into the open-loop system:

$$\begin{aligned} G(s) &= R(s) \cdot P^{-1}(s) \cdot E \\ &= T(s) \cdot E \end{aligned}$$

The total transfer function can be realized by a simple series compensation:

$$G_{tot}(s) = T(s) \cdot G_c(s) \cdot E$$



- Obviously, any linear state feedback can be realized as a series compensation with a proper transfer function  $G_c(s)$ .
- This is not desirable since we lose all the nice properties of feedback control (reduced sensitivity), and since we cancel unstable open-loop poles with unstable zeros which is utterly dubious.

Question: Can we do the opposite, i.e., can we first design a series compensator (which is easy), and then try to realize as large a part of this compensator as possible through an equivalent feedback loop?

## "Division" of PM's :

Analogy:  $7/3 = \underbrace{2.333\dots}_{\text{not integer}}$

We can write :

$$7 = \underbrace{2 \cdot 3}_{\text{divisor}} + \underbrace{1}_{\text{remainder}}$$

⇒ Given two polynomial matrices:

$$J(s) \in \mathbb{R}_p^{p \times m} \quad \text{and} \quad \bar{J}(s) \in \mathbb{R}_p^{q \times m}$$

where  $\mathcal{S}(J(s)) = m$

⇒ It is always possible to find two polynomial matrices:

$$M(s) \in \mathbb{R}_p^{q \times p} \quad \text{and} \quad N(s) \in \mathbb{R}_p^{q \times m}$$

such that:

$$\bar{J}(s) = \underbrace{M(s)}_{\text{divisor}} \cdot J(s) + \underbrace{N(s)}_{\text{remainder}}$$

Algorithm:

We define  $\hat{J}(s)$  to be a  $m \times m$  Polynomial matrix consisting of  $m$  linearly independent rows of  $J(s)$

$\Rightarrow \bar{J}(s) \cdot \hat{J}^{-1}(s)$  is then a rational transfer function matrix (which however does not need to be proper)

$$\bar{J}(s) \cdot \hat{J}^{-1}(s) \in \mathbb{R}_G^{q \times m}$$

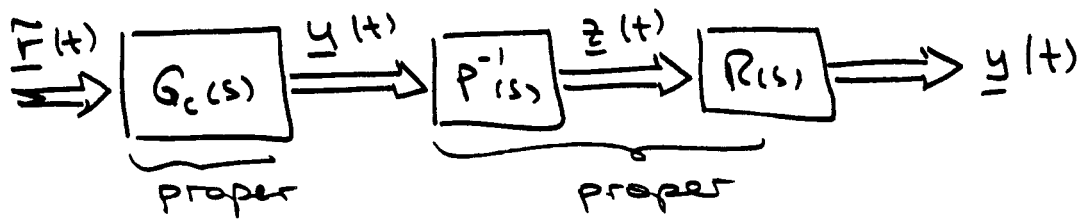
We divide through:

$$\bar{J}(s) \cdot \hat{J}^{-1}(s) = \hat{M}(s) + \underbrace{N(s) \cdot \hat{J}^{-1}(s)}_{\text{strictly proper}}$$

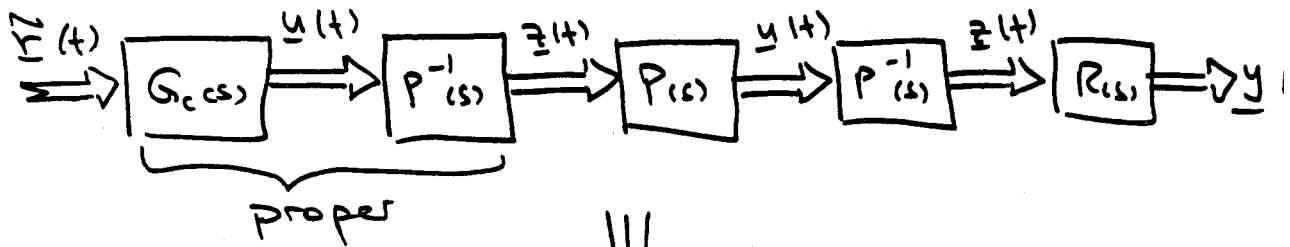
$$\Rightarrow \bar{J}(s) = \hat{M}(s) \cdot \hat{J}(s) + N(s)$$

$\Rightarrow$  Fill the eliminated rows of  $J$  back in, and add "zero"-columns in  $\hat{M}(s)$

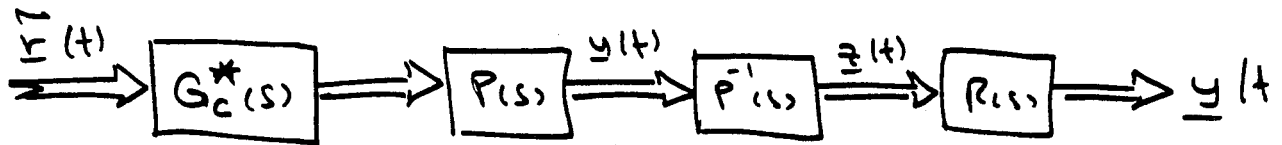
$$\Rightarrow \bar{J}(s) = M(s) \cdot J(s) + N(s)$$



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$$G_c^*(s) = P^{-1}(s) \cdot G_c(s)$$

We regroup terms, and decompose  $G_c^*(s)$  as:

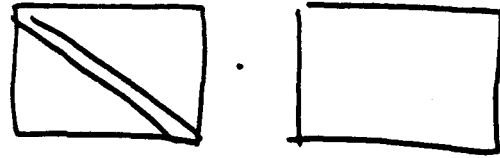
$$G_c^*(s) = P_c^{-1}(s) \cdot L_c(s)$$

where:  $\{P_c(s), L_c(s)\}$  are relative left prime (e.g. observer - canonical decomposition). Furthermore, we get  $L_c(s)$  into upper-triangular form.

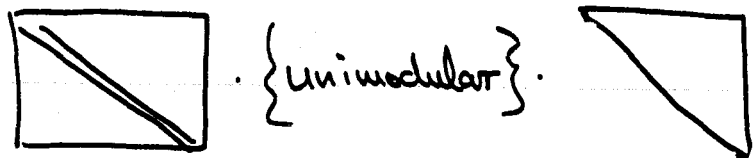
Algorithm:

$$G_c^*(s) = P_c^{*-1}(s) \cdot L_c^*(s)$$

observer -  
canonical  
decomposition



$$\Rightarrow G_c^*(s) = P_c^{*-1}(s) \cdot U_L(s) \cdot L_c(s)$$

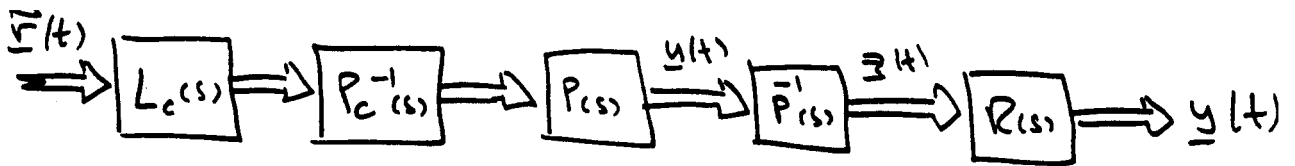


$$= P_c^{*-1}(s) \cdot (U_L^{-1}(s))^{-1} \cdot L_c(s)$$

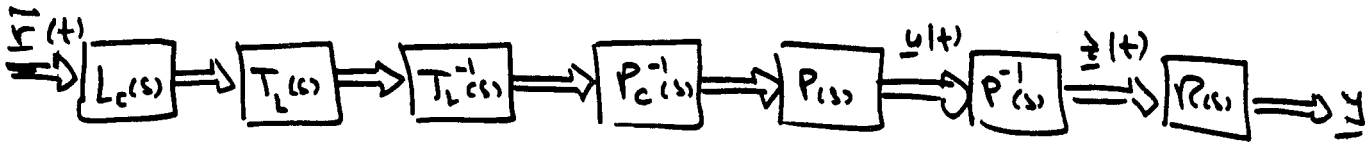
$$= \underbrace{[U_L^{-1}(s) \cdot P_c^*(s)]^{-1}}_{P_c(s)} \cdot L_c(s)$$

$$\Rightarrow G_c^*(s) = P_c(s) \cdot L_c(s)$$





We expand once more:

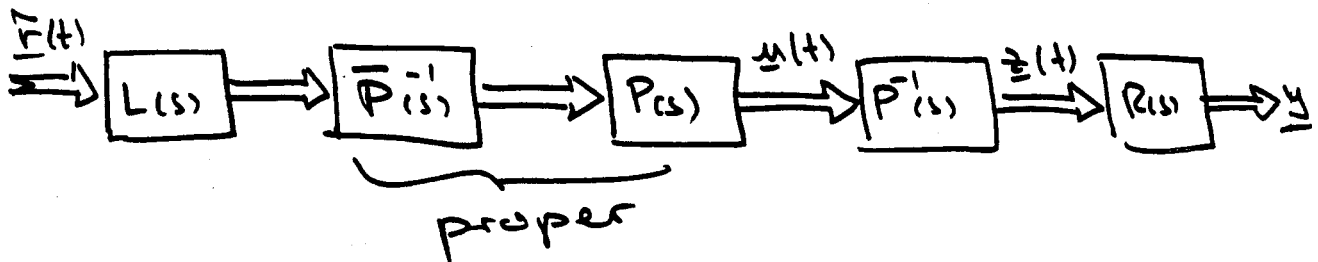


We choose  $T_L(s)$  such that:

- Roots  $\xi | T_L(s) | \xi$  are stable
- $P(s) \cdot P_c^{-1}(s) \cdot T_L^{-1}(s)$  is proper

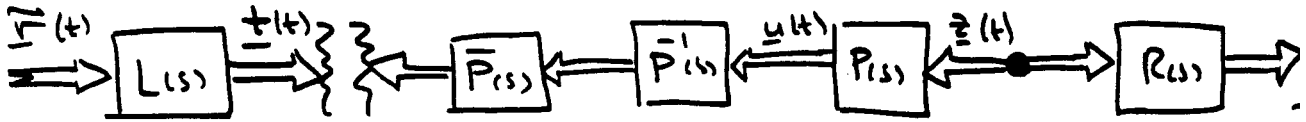
If  $P(s) \cdot P_c^{-1}(s)$  is already proper  
 $\Rightarrow$  omit the step ( $T_L(s) = I$ ).

It is always possible to make  $T_L(s)$  diagonal.



$$\left| \begin{array}{l} L(s) = T_L(s) \cdot L_c(s) \\ \overline{P}(s) = T_L(s) \cdot P_c(s) \end{array} \right|$$

We redraw:

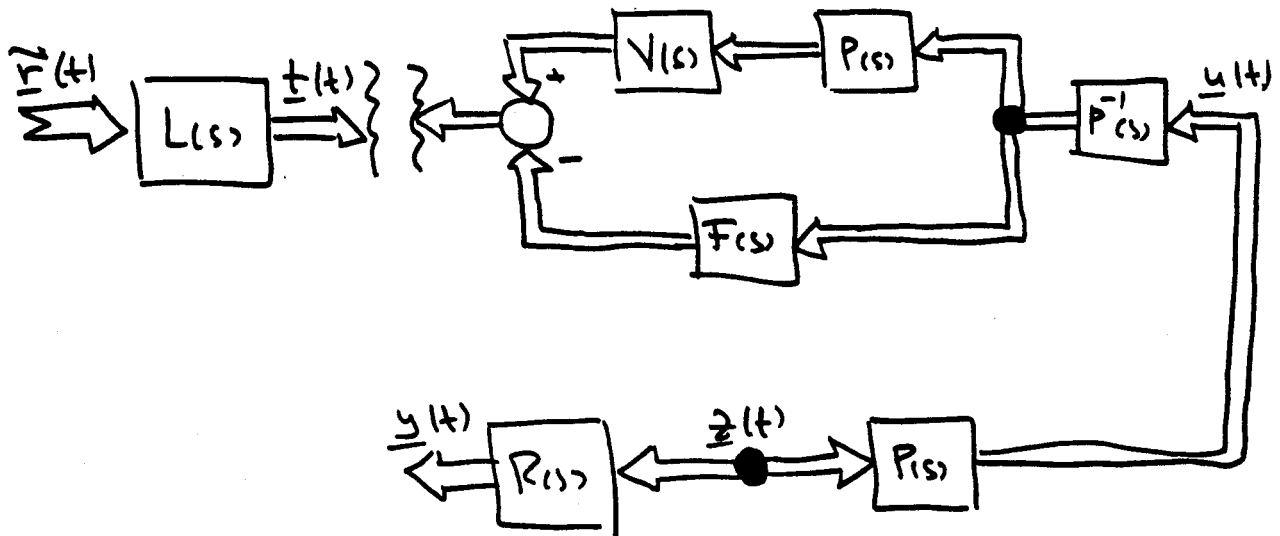


We use the division algorithm:

$$\bar{P}(s) = V(s) \cdot P(s) - F(s)$$

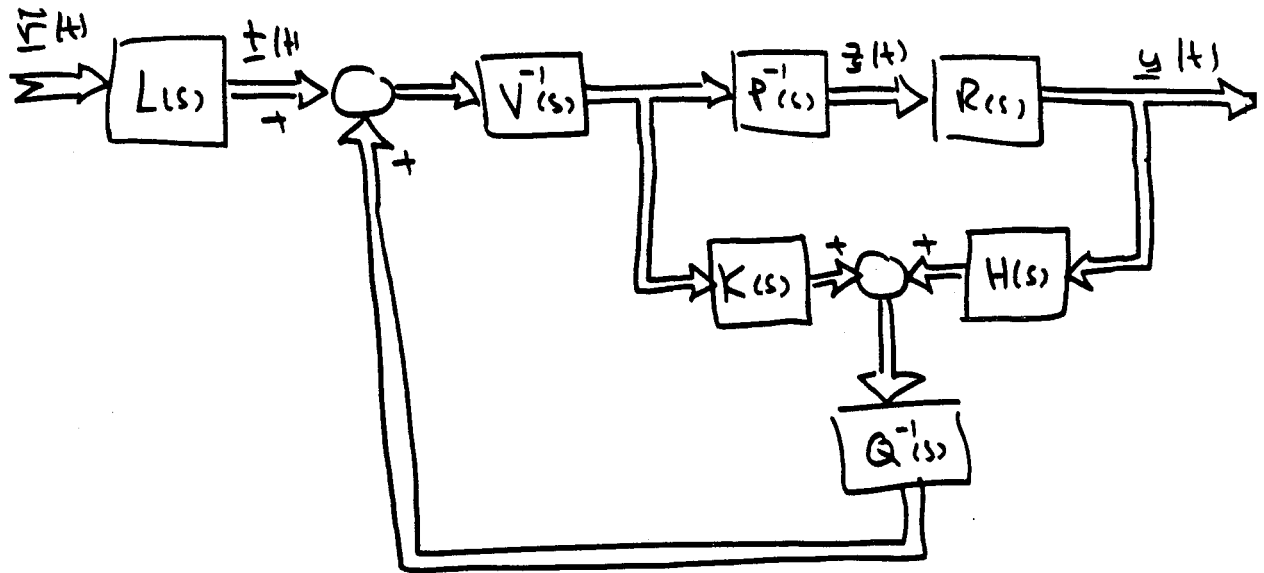
where:  $V(s), F(s) \in \mathbb{R}_p^{m \times m}$

$$\partial_c [F(s)] < \partial_c [P(s)]$$



Let us compare what we got with what we wish to get:





$$\Rightarrow \underline{z}(t) = [V(s) \cdot P(s) - F(s)] \underline{z}(s)$$

$$\equiv \left[ V(s) \cdot P(s) - Q^{-1}(s) (K(s) \cdot P(s) + H(s) \cdot R(s)) \right] \cdot \underline{z}(s)$$

$$\Rightarrow \boxed{Q(s) \cdot F(s) = B(s) = K(s) P(s) + H(s) R(s)}$$

We already showed how to find  $\{Q(s), K(s), H(s)\}$  from this equation.

Example: Given the system

$$G(s) = \begin{bmatrix} \frac{s+1}{s^2} & \frac{s+2}{s^2+1} \\ \frac{2}{s} & \frac{2s+3}{s^2+1} \end{bmatrix} = \underbrace{\begin{bmatrix} (s+1) & 1 \\ 2s & 3 \end{bmatrix}}_{R(s)} \cdot \underbrace{\begin{bmatrix} s & -s^2 \\ \phi & (s^2+1) \end{bmatrix}}_{P(s)}$$

$\{R(s), P(s)\}$  are relative right prime.

We want:

$$G_{tot}(s) = \begin{bmatrix} \frac{1}{s^2+3s+2} & \phi \\ \phi & \frac{1}{s^2+3s+2} \end{bmatrix}$$

$$\Rightarrow G_{tot}(s) = G(s) \cdot G_c(s) = R(s) \cdot P^{-1}(s) \cdot G_c(s)$$

$$\Rightarrow G_c(s) = P(s) \cdot R^{-1}(s) \cdot G_{tot}(s)$$

$$\Rightarrow G_c(s) = \begin{bmatrix} \frac{2s^3+3s^2}{s^3+6s^2+11s+6} & \frac{-s^3-2s^2}{s^3+6s^2+11s+6} \\ \frac{-2s^3-2s}{s^3+6s^2+11s+6} & \frac{s^3+s^2+s+1}{s^3+6s^2+11s+6} \end{bmatrix}$$

Is the series compensator.

$$(i) \quad P^{-1}(s) \cdot G_c(s) = \frac{-536 - \begin{bmatrix} 3 & -1 \\ -2s & (s+1) \end{bmatrix}}{s^3 + 6s^2 + 11s + 6}$$

$$\equiv P_c^{-1}(s) \cdot L_c(s) = \underbrace{\begin{bmatrix} (s^3 + 6s^2 + 11s + 6) & \Phi \\ (2s^3 + 6s^2 + 4s) & (3s^2 + 9s + 6) \end{bmatrix}^{-1}}_{P_c(s)} \cdot \underbrace{\begin{bmatrix} 3 \\ \Phi \end{bmatrix}}_{L_c(s)}$$

$L_c(s) = \begin{bmatrix} 3 \\ \Phi \end{bmatrix}$ ;  $\{P_c(s), L_c(s)\}$  are relative left prime.

(ii)  $P(s) \cdot P_c^{-1}(s)$  is proper  $\Rightarrow T_L(s) = I$

$$\Rightarrow \overline{P}(s) \equiv P_c(s) \quad ; \quad L(s) \equiv L_c(s)$$

$$(iii) \quad \overline{P}(s) \cdot P^{-1}(s) = \underbrace{\begin{bmatrix} (s+6) & (s+6) \\ (2s+6) & (2s+9) \end{bmatrix}}_{V(s)} - \underbrace{\begin{bmatrix} \frac{-11s-6}{s^2} & \frac{-10s}{s^2+1} \\ \frac{-4}{s} & \frac{-11s+3}{s^2+1} \end{bmatrix}}_{F(s) \cdot P^{-1}(s)}$$

$$= \begin{bmatrix} (-11s-6) & (s+6) \\ -4s & (-7s+3) \end{bmatrix}$$

$$V^{-1}(s) = \begin{bmatrix} \frac{\frac{2}{3}s+3}{s+6} & \frac{-1}{s} \\ \frac{-\frac{2}{3}s-2}{s+6} & \frac{1}{s} \end{bmatrix} \quad ; \quad L(s) = \begin{bmatrix} 3 & -1 \\ \Phi & 1 \end{bmatrix}$$

$\Rightarrow$  Find  $\{ H(s), K(s), Q(s) \}$  according to the last chapter, where  $\partial[|Q(s)|] = 2$ .

## Decoupling and Model Reduction

### "Inversion" of Dynamical Systems:

Def:  $G_{LI}(s)$  is called left-inverse to  $G(s)$  iff:

$$G_{LI}(s) \cdot G(s) = I^{(m)}$$

Def:  $G_{RI}(s)$  is called right-inverse to  $G(s)$  iff:

$$G(s) \cdot G_{RI}(s) = I^{(p)}$$

Theorem: • A system has only a left-inverse if

$$\mathcal{R}[G(s)] = m$$

• A system has only a right-inverse if

$$\mathcal{R}[G(s)] = p$$

Note: • In general,  $G_{LI}(s)$  and  $G_{RI}(s)$  are not proper.

• If  $m=p$ , and  $G(s)$  is non-singular

$$\Rightarrow G_{LI}(s) = G_{RI}(s) = G^{-1}(s)$$

• If  $m < p \Rightarrow G_{LI}(s)$  is not unique  
 $G_{RI}(s)$  does not exist

• If  $m > p \Rightarrow G_{RI}(s)$  does not exist  
 $G_{LI}(s)$  is not unique.

### Decoupling:

Theorem: A system  $G(s) \in \mathbb{R}_G^{p \times m}$  can be decoupled only if it is right-invertible, i.e.  
$$P \equiv \mathcal{S}[G(s)]$$

Proof:

(a) We find  $G_{RI}(s)$  such that:

$$G(s) \cdot G_{RI}(s) = I^{(p)}$$

Then we construct:

$$G_c(s) = G_{RI}(s) \cdot [\text{diag}\{p_i(s)\}]^{-1}$$

by choosing  $p_i(s)$  such that  $G_c(s)$  is proper.

$\Rightarrow G_c(s)$  is a series-compensator which decouples the system

$$G_{tot}(s) = G(s) \cdot G_c(s) = G(s) \cdot G_{RI}(s) \cdot [\text{diag}\{p_i(s)\}]^{-1} \equiv [\text{diag}\{p_i(s)\}]^{-1}$$

q.e.d.

(b) If  $G(s)$  can be decoupled with  $G_c(s)$

$$\Rightarrow G(s) \cdot G_c(s) = G_d(s)$$

$\uparrow$  diagonal, non-singular

- 540 -

$$\Rightarrow G(s) \cdot \underbrace{G_c(s) \cdot G_d^{-1}(s)}_{G_{RT}(s)} = I^{(p)}$$

q.e.d.

Warnings: We obtain immediately an algorithm for dynamical decoupling. However, there remain a number of questions to be answered:

- (1) How many poles can be chosen freely beside from decoupling?
- (2) Are there invariant poles?
- (3) What is the minimal order of the decoupling compensator?
- (4) Can state-feedback alone decouple the system?

- Answers:
- There exist already algorithms to design compensators which minimize the dynamics and allow to move as many poles as possible (non-trivial!)
  - There can indeed exist invariant poles which appear in the decoupled system as unobservable modes. These must (of course) be stable.
  - These algorithms compute the minimal order of the compensator.



## Static Decoupling:

A somewhat simpler problem is the static decoupling problem.

Here, we do not request that one input influences only one output at all times, but that, in the steady-state, a change in one input has only a permanent effect on not more than one output, and that every output is affected by exactly one input.

Usually:

$$\lim_{s \rightarrow 0} G_{tot}(s) = I \quad (p \times m)$$

$$\text{e.g. } G_{tot}(s) = \begin{bmatrix} 1 & \phi & \phi & \phi \\ \phi & 1 & \phi & \phi \\ \phi & \phi & 1 & \phi \end{bmatrix}$$

Obviously, this is only possible if  $m \geq p$ .

Def:  $G_{tot}(s)$  is statically decoupled  
iff:

- (i)  $G_{tot}(s)$  is asymptotically stable
- (ii)  $\lim_{s \rightarrow \infty} G_{tot}(s)$  is diagonal, non-singular

Algorithm:

We use state feedback:

$$\underline{y}(t) = \underline{F} \cdot \underline{x}(t) + \underline{E} \cdot \underline{r}(t)$$

$$\Rightarrow G_{tot}(s) = \underline{R}(s) [\underline{P}(s) - \underline{F}(s)]^{-1} \cdot \underline{E}$$

Choose  $\underline{F}$  such that:

- $|\underline{P}(s) - \underline{F}(s)|$  has all poles in the left half plane
- $|\underline{P}(\infty) - \underline{F}(\infty)|$  is non-singular

Choose  $\underline{E}$  such that:

$$\begin{aligned} \lim_{s \rightarrow \infty} G_{tot}(s) &= \underline{R}(\infty) [\underline{P}(\infty) - \underline{F}(\infty)]^{-1} \cdot \underline{E} \\ &= \underline{\Lambda}, \text{ diagonal } \in \mathbb{R}^{p \times p} \end{aligned}$$

Obviously, this is only possible if:

$$\mathcal{S}\{R(\omega)\} = p$$

Theorem: The static decoupling problem is solvable if

$$\mathcal{S}\{R(\omega)\} = p.$$

Note: •  $F$  is only used for stabilization. Static decoupling is achieved through the choice of  $E$ .

• Obviously, static decoupling is only possible if:

$$m \geq p$$

- If  $m = p \Rightarrow E$  is unique
- If  $m > p \Rightarrow E$  is not unique.

## Model Reconstruction:

Problem: Given a system with  $G(s)$  and a model of that system:  $G_m(s)$ . We would like to find  $G_c(s)$  such that:

$$G(s) \cdot G_c(s) \equiv G_{tot}(s) \equiv G_m(s)$$

Two cases:

(a)  $p \geq m$ ,  $\text{rank}[G(s)] = m$   
 $\Rightarrow G(s)$  is left-invertible.

$$\begin{matrix} m \\ \boxed{G(s)} \\ p \end{matrix} \cdot \begin{matrix} q \\ \boxed{G_c(s)} \\ m \end{matrix} = \begin{matrix} q \\ \boxed{G_m(s)} \\ p \end{matrix}$$

Algorithm:

We construct  $\hat{G}(s)$  from any  $m$  linearly independent rows of  $G(s)$ .

We construct  $\hat{G}_m(s)$  from the same rows of  $G_m(s)$ .

$$\Rightarrow \hat{G}(s) \cdot G_c(s) = \hat{G}_m(s)$$

$$\Rightarrow G_c(s) = \hat{G}^{-1}(s) \cdot \hat{G}_m(s)$$

$$\Rightarrow G_m(s) = G(s) \cdot G_c(s) = G(s) \cdot \hat{G}^{-1}(s) \cdot \hat{G}_m(s)$$

$\Rightarrow$  We can choose only  $m$  rows of  $G_m(s)$  freely. A further restriction comes from the fact that  $G_c(s)$  should be proper.

$$(b) \quad m \geq p, \quad \mathcal{R}[G(s)] = p$$

$\Rightarrow G(s)$  is right-invertible

$$\begin{array}{c}
 m \\
 \boxed{G(s)} \\
 p
 \end{array}
 \cdot
 \begin{array}{c}
 q \\
 \boxed{G_c(s)} \\
 m
 \end{array}
 =
 \begin{array}{c}
 q \\
 \boxed{G_m(s)} \\
 p
 \end{array}$$

Same algorithm as before. However, here the solution is not unique. We can use the additional freedom to:

- (i) ensure that  $G_c(s)$  is proper
- (ii) the compensated system has stable poles only
- (iii) the additional system order is minimal.